

ECE 2020

Stock price prediction using Convolutional Neural Network



Member

1. Thịnh – Leader

2. Nam – Data collection

3. Hoàng – Model design & training

4. Trí – Made Powerpoint



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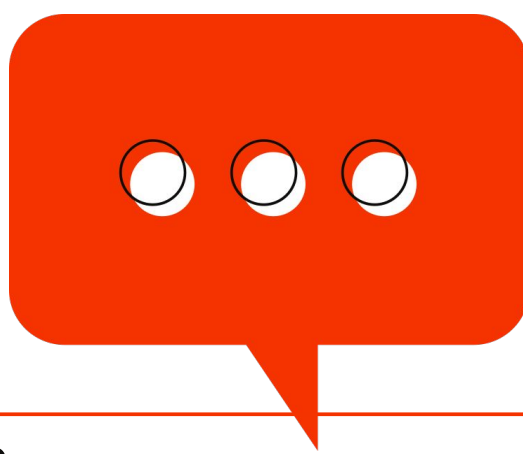
Introduction



What is stock?

Stock is a small fraction of a company. Stocks can be traded using money.

For example, FPT Telecom company stock cost 80600VND per stock.



**Stock price can go up
and down**
→ **Gain huge potential
if we can predict**
→ **Use Convolutional
Neural Network**



Target



1

Produce a model that can predict
FPT stock price of the next day by
using close price of the last 50 days.



2 Analyze its accuracy.

Related



stock price prediction using convolutional neural networks



<https://iopscience.iop.org> › article › pdf

Stock Prediction Using Convolutional Neural Network

viết bởi S Chen · 2018 · Trích dẫn 82 bài viết — In this paper, we proposed a deep learning method based on **Convolutional Neural Network** to **predict** the **stock price** movement of...

<https://arxiv.org> › q-fin ▼ Dịch trang này

Stock Price Prediction Using Convolutional Neural Networks ...

viết bởi S Mehtab · 2020 · Trích dẫn 92 bài viết — In this work, we propose a **hybrid approach** for stock price prediction using machine learning and deep learning-based methods.

<https://www.researchgate.net> › ... › Stocks · Dịch trang này

(PDF) Stock Prediction Using Convolutional Neural Network

In this paper, we proposed a **deep learning method based** on Convolutional Neural Network to predict the stock price movement of Chinese stock market.



Data Preparation – 6 steps

Step 1



A	B	C	D	E	F
ficker>	<DTYYYYMMDD>	<OpenFixed>	<HighFixed>	<LowFixe	
»T,20061213,	15.6977,	15.6977,	15.6977,	15.6977,	83530
»T,20061214,	16.4826,	16.4826,	16.4826,	16.4826,	280710
»T,20061215,	17.3067,	17.3067,	17.3067,	17.3067,	265300
»T,20061218,	18.1701,	18.1701,	18.1701,	18.1701,	215790
»T,20061219,	19.0727,	19.0727,	19.0727,	19.0727,	137520
»T,20061220,	20.0146,	20.0146,	20.0146,	20.0146,	220120
»T,20061221,	20.0146,	20.0146,	19.0335,	19.0335,	156400
»T,20061222,	18.0916,	18.0916,	18.0916,	18.0916,	143500
»T,20061225,	17.1890,	17.1890,	17.1890,	17.1890,	22140
»T,20061226,	18.0131,	18.0131,	16.3649,	18.0131,	300150
»T,20061227,	18.0131,	18.8373,	18.0131,	18.8373,	413230
»T,20061228,	18.8373,	18.8373,	18.4449,	18.8373,	144840
»T,20061229,	18.2486,	18.2486,	18.0524,	18.0524,	69760
»T,20070102,	17.2675,	17.6600,	17.2675,	17.2675,	81170
»T,20070103,	17.6599,	17.6599,	17.2675,	17.6599,	74660
»T,20070104,	18.5233,	18.5233,	18.5233,	18.5233,	148510
»T,20070105,	19.4259,	19.4259,	19.4259,	19.4259,	377800
»T,20070108,	20.0146,	20.0146,	19.2297,	19.2297,	351470
»T,20070109,	19.4259,	19.4259,	19.2297,	19.2297,	225270
»T,20070110,	19.2297,	19.4259,	19.2297,	19.4259,	314250
»T,20070111,	19.6221,	19.6221,	19.6221,	19.6221,	283920
»T,20070112,	20.6032,	20.6033,	20.6032,	20.6033,	279710
»T,20070115,	20.9957,	21.6237,	20.9957,	21.6236,	274290
»T,20070116,	22.6832,	22.6832,	22.6832,	22.6832,	339760
»T,20070117,	23.5073,	23.5073,	21.5844,	21.5844,	145870
»T,20070118,	20.5248,	21.6236,	20.5248,	21.6236,	282430
»T,20070119,	22.6832,	22.6832,	22.6832,	22.6832,	95290
»T,20070122,	23.7820,	23.7820,	23.5466,	23.5466,	461630
»T,20070123,	23.5466,	23.6251,	23.5466,	23.6251,	243990

Download the Financial Statement of
FPT.

<https://www.cophieu68.vn/export.php>



Step 2

The file was downloaded in.csv format; therefore, we must use a module called pandas to convert it to.xlsx format (regular Excel).

```
import pandas as pd

read_file = pd.read_csv(r'C:/Users/luciu/PycharmProjects/MCT1/fpt.csv')
read_file.to_excel(r'C:/Users/luciu/PycharmProjects/MCT1/fpt.xlsx', index = None, header=True)
```

	A	B	C	D	E	F
1	<Ticker>,<DTYYYYMMDD>,<OpenFixed>,<HighFixed>,<LowFixed>,<Volume>					
2	FPT	20061213	15.6977	15.6977	15.6977	83530
3	FPT	20061214	16.4826	16.4826	16.4826	280710
4	FPT	20061215	17.3067	17.3067	17.3067	265300
5	FPT	20061218	18.1701	18.1701	18.1701	215790
6	FPT	20061219	19.0727	19.0727	19.0727	137520
7	FPT	20061220	20.0146	20.0146	20.0146	220120
8	FPT	20061221	20.0146	20.0146	19.0335	156400
9	FPT	20061222	18.0916	18.0916	18.0916	143500
10	FPT	20061225	17.1890	17.1890	17.1890	22140
11	FPT	20061226	18.0131	18.0131	16.3649	300150
12	FPT	20061227	18.0131	18.8373	18.0131	413230
13	FPT	20061228	18.8373	18.8373	18.4449	144840
14	FPT	20061229	18.2486	18.2486	18.0524	69760
15	FPT	20070102	17.2675	17.6600	17.2675	81170
16	FPT	20070103	17.6599	17.6599	17.2675	74660
17	FPT	20070104	18.5233	18.5233	18.5233	148510
18	FPT	20070105	19.4259	19.4259	19.4259	377800
19	FPT	20070108	20.0146	20.0146	19.2297	351470
20	FPT	20070109	19.4259	19.4259	19.2297	225270
21	FPT	20070110	19.2297	19.4259	19.2297	314250
22	FPT	20070111	19.6221	19.6221	19.6221	283920
23	FPT	20070112	20.6032	20.6033	20.6032	279710
24	FPT	20070115	20.9957	21.6237	20.9957	274290
25	FPT	20070116	22.6832	22.6832	22.6832	339760
26	FPT	20070117	23.5073	23.5073	21.5844	145870
27	FPT	20070118	20.5248	21.6236	20.5248	282430
28	FPT	20070119	22.6832	22.6832	22.6832	95290
29	FPT	20070122	23.7820	23.7820	23.5466	461630

csv file

	A	B	C	D	E	F	G	H
1	<Ticker>,<DTYYYYMMDD>,<OpenFixed>,<HighFixed>,<LowFixed>,<CloseFixed>,<Volume>							
2	FPT	20061213	15.6977	15.6977	15.6977	15.6977	83530	
3	FPT	20061214	16.4826	16.4826	16.4826	16.4826	280710	
4	FPT	20061215	17.3067	17.3067	17.3067	17.3067	265300	
5	FPT	20061218	18.1701	18.1701	18.1701	18.1701	215790	
6	FPT	20061219	19.0727	19.0727	19.0727	19.0727	137520	
7	FPT	20061220	20.0146	20.0146	20.0146	20.0146	220120	
8	FPT	20061221	20.0146	20.0146	19.0335	19.0335	156400	
9	FPT	20061222	18.0916	18.0916	18.0916	18.0916	143500	
10	FPT	20061225	17.189	17.189	17.189	17.189	22140	
11	FPT	20061226	18.0131	18.0131	16.3649	18.0131	300150	
12	FPT	20061227	18.0131	18.8373	18.0131	18.8373	413230	
13	FPT	20061228	18.8373	18.8373	18.4449	18.8373	144840	
14	FPT	20061229	18.2486	18.2486	18.0524	18.0524	69760	
15	FPT	20070102	17.2675	17.66	17.2675	17.2675	81170	
16	FPT	20070103	17.6599	17.6599	17.2675	17.6599	74660	
17	FPT	20070104	18.5233	18.5233	18.5233	18.5233	148510	
18	FPT	20070105	19.4259	19.4259	19.4259	19.4259	377800	
19	FPT	20070108	20.0146	20.0146	19.2297	19.2297	351470	
20	FPT	20070109	19.4259	19.4259	19.2297	19.2297	225270	
21	FPT	20070110	19.2297	19.4259	19.2297	19.4259	314250	
22	FPT	20070111	19.6221	19.6221	19.6221	19.6221	283920	
23	FPT	20070112	20.6032	20.6033	20.6032	20.6033	279710	
24	FPT	20070115	20.9957	21.6237	20.9957	21.6236	274290	
25	FPT	20070116	22.6832	22.6832	22.6832	22.6832	339760	
26	FPT	20070117	23.5073	23.5073	21.5844	21.5844	145870	
27	FPT	20070118	20.5248	21.6236	20.5248	21.6236	282430	
28	FPT	20070119	22.6832	22.6832	22.6832	22.6832	95290	
29	FPT	20070122	23.782	23.782	23.5466	23.5466	461630	
30	FPT	20070123	23.5466	23.6251	23.5466	23.6251	243990	
31	FPT	20070124	23.6251	23.6251	22.7617	22.7617	156420	
32	FPT	20070125	22.7617	22.7617	22.7617	22.7617	142420	

xlsx file



Step 3

After that, we create another excel file that contains the data that we want to train, here is the closing price

7	20170110	16.1421
8	20170111	16.1772
9	20170112	16.1246
10	20170113	15.7561
11	20170116	15.4403
12	20170117	15.5806
13	20170118	15.3701
14	20170119	15.3701
15	20170120	15.5105
16	20170123	15.5982
17	20170124	15.6859
18	20170125	15.8263
19	20170202	15.7561



Step 4

Because we want to use the sliding window approach to construct the dataset.

For example, the data will be compiled:

[Day 1] to [Day 50] and [Day 51] is increase/decrease

[Day 2] to [Day 51] and [Day 52] is increase/decrease

...

[Day 250] to [Day 299] and [Day 300] is increase/decrease

We shall put 1 at the end of the row if the day after increases; else, we

```
15.9316,16.0719,16.1421,16.1421,16.3351,16.0895,16.0368,15.9491,16.0193,16.1246,16.0719,16.0719,15.9667,16.1421,16.0544,16.0368,15.9842,16.0368,16.107,16.3527,-1
15.9316,16.0719,16.1421,16.1421,16.3351,16.0895,16.0368,15.9491,16.0193,16.1246,16.0719,16.0719,15.9667,16.1421,16.0544,16.0368,15.9842,16.0368,16.107,16.3527,-1
16.0719,16.1421,16.1421,16.3351,16.0895,16.0368,15.9491,16.0193,16.1246,16.0719,16.0719,15.9667,16.1421,16.0544,16.0368,15.9842,16.0368,16.107,16.3527,16.3527,-1
16.1421,16.1421,16.3351,16.0895,16.0368,15.9491,16.0193,16.1246,16.0719,16.0719,15.9667,16.1421,16.0544,16.0368,15.9842,16.0368,16.107,16.3527,16.3527,16.1772,1
16.1421,16.3351,16.0895,16.0368,15.9491,16.0193,16.1246,16.0719,16.0719,15.9667,16.1421,16.0544,16.0368,15.9842,16.0368,16.107,16.3527,16.3527,16.1772,16.8439,-1
```


Step 5

01

After that, we import the openpyxl module to use an excel file.

```
import openpyxl as op
```

02

Then we set some basic data of the excel file.

```
wb = op.load_workbook("s.xlsx")
sh = wb.active
col = 2
num_rows = 40050
start = 1
row = 0
row_i = 2
count = 1
file = open("C:/Users/luciu/PycharmProjects/MCT1/s.txt", 'w')
```

03

When we write in a text file, the letter "w" at the end indicates that we are overwriting the text.

```
file = open("C:/Users/luciu/PycharmProjects/MCT1/s.txt", 'w')
```

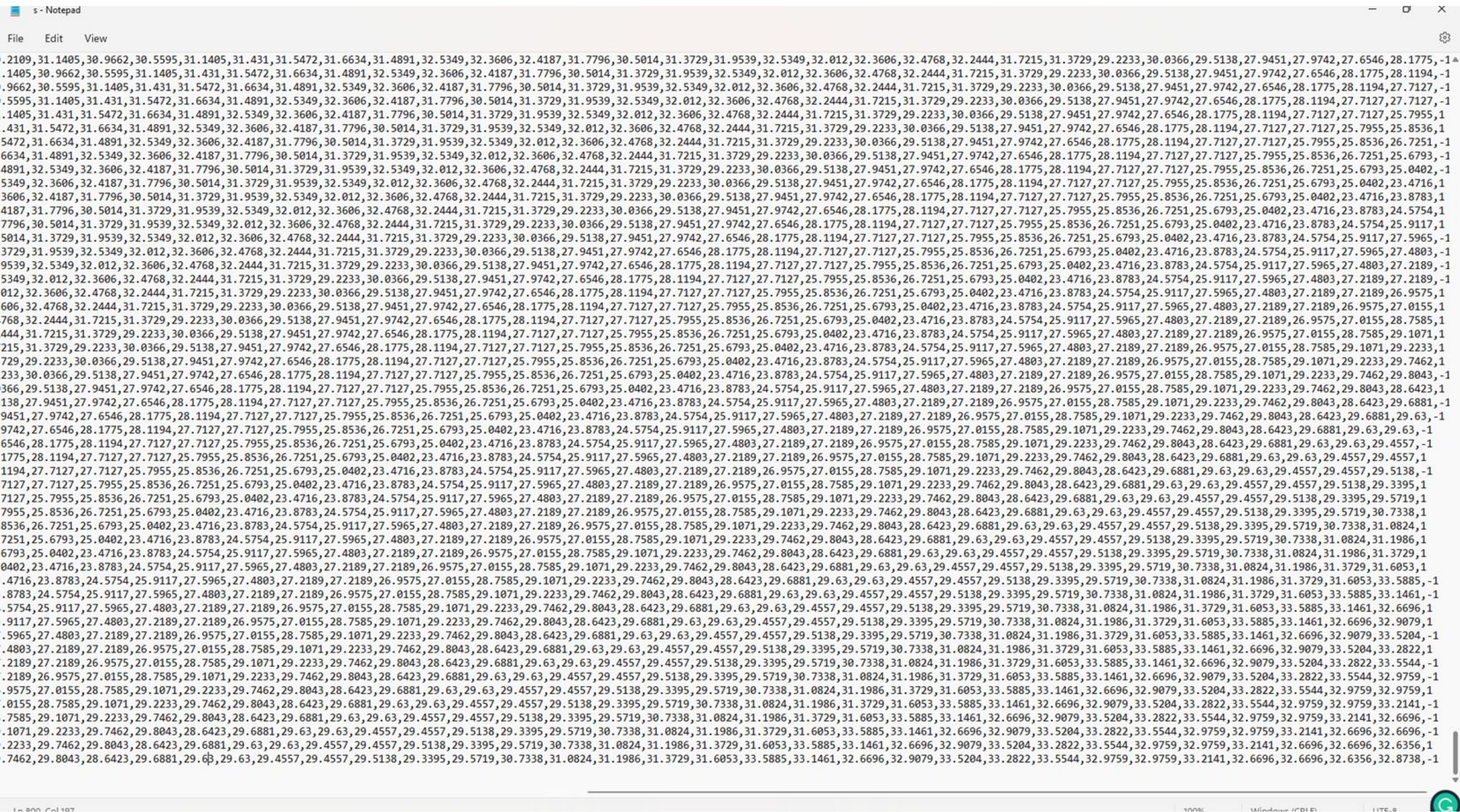


Step 6

Then, using a loop to separate each value from the others using commas, we write each value into the text file. When there are 50 values, it will compare the number from the previous day with the next day, and if the difference is greater, it will write -1; otherwise, it will write 1. The next line's starting value will be the value for the following day as we restart the counting. If there are not enough values to make 50, it will keep counting.

```
while row < num_rows+1:
    file.write(str(sh.cell(row_i, col).value)+ ",")
    print(row)
    if count == 50:
        if sh.cell(row_i, col).value >= sh.cell(row_i +1, col).value:
            file.write('-1')
        else:
            file.write('1')

        start = start + 1
        row_i = start
        count = 1
        file.write('\n')
    else:
        row_i +=1
        count +=1
    row += 1
```

The
result



Model training

✓
0
giây

```
[1] from google.colab import drive  
import pandas as pd  
import numpy as np
```

✓
26
giây

```
[2] drive.mount('/content/gdrive')
```

Mounted at /content/gdrive

Connect to GG drive

```
[3] train=pd.read_csv(r'/content/gdrive/MyDrive/SSA_lab/train.txt', header=None)  
test=pd.read_csv(r'/content/gdrive/MyDrive/SSA_lab/test.txt', header=None)
```



✓
0
giây

[4] train

	0	1	2	3	4	5	6	7	8	9	...	41
0	15.7386	15.8789	15.8263	15.7737	15.8614	16.1421	16.1772	16.1246	15.7561	15.4403	...	16.0719
1	15.8789	15.8263	15.7737	15.8614	16.1421	16.1772	16.1246	15.7561	15.4403	15.5806	...	15.9667
2	15.8263	15.7737	15.8614	16.1421	16.1772	16.1246	15.7561	15.4403	15.5806	15.3701	...	16.1421
3	15.7737	15.8614	16.1421	16.1772	16.1246	15.7561	15.4403	15.5806	15.3701	15.3701	...	16.0544
4	15.8614	16.1421	16.1772	16.1246	15.7561	15.4403	15.5806	15.3701	15.3701	15.5105	...	16.0368
...
795	28.1775	28.1194	27.7127	27.7127	25.7955	25.8536	26.7251	25.6793	25.0402	23.4716	...	32.9079
796	28.1194	27.7127	27.7127	25.7955	25.8536	26.7251	25.6793	25.0402	23.4716	23.8783	...	33.5204
797	27.7127	27.7127	25.7955	25.8536	26.7251	25.6793	25.0402	23.4716	23.8783	24.5754	...	33.2822
798	27.7127	25.7955	25.8536	26.7251	25.6793	25.0402	23.4716	23.8783	24.5754	25.9117	...	33.5544
799	25.7955	25.8536	26.7251	25.6793	25.0402	23.4716	23.8783	24.5754	25.9117	27.5965	...	32.9759

800 rows × 51 columns



✓
0
giây

[5] test

	0	1	2	3	4	5	6	7	8	9	...	41
0	33.5885	33.3162	33.2141	31.3084	31.9890	31.3084	31.7168	31.5466	31.7168	31.9890	...	31.6827
1	33.3162	33.2141	31.3084	31.9890	31.3084	31.7168	31.5466	31.7168	31.9890	31.9890	...	31.9890
2	33.2141	31.3084	31.9890	31.3084	31.7168	31.5466	31.7168	31.9890	31.9890	32.3634	...	31.8529
3	31.3084	31.9890	31.3084	31.7168	31.5466	31.7168	31.9890	31.9890	32.3634	32.1251	...	32.0571
4	31.9890	31.3084	31.7168	31.5466	31.7168	31.9890	31.9890	32.3634	32.1251	31.8529	...	32.1251
...
195	56.1708	56.3098	55.4756	54.7109	54.7804	53.5291	53.2510	52.9034	53.9462	53.5986	...	60.4809
196	56.3098	55.4756	54.7109	54.7804	53.5291	53.2510	52.9034	53.9462	53.5986	54.0157	...	63.1226
197	55.4756	54.7109	54.7804	53.5291	53.2510	52.9034	53.9462	53.5986	54.0157	55.4061	...	66.3900
198	54.7109	54.7804	53.5291	53.2510	52.9034	53.9462	53.5986	54.0157	55.4061	55.4756	...	65.2081
199	54.7804	53.5291	53.2510	52.9034	53.9462	53.5986	54.0157	55.4061	55.4756	55.3365	...	64.9996

200 rows × 51 columns



Importing dataset to Python

✓
0
giây

```
[6] xtrain=train.iloc[:,0:50]  
    xtrain=np.array(xtrain)  
    xtrain=np.reshape(xtrain,(800,50))
```

✓
0
giây

```
[7] ytrain=train.iloc[:,50:51]  
    ytrain=np.array(ytrain)  
    ytrain=np.reshape(ytrain,(800,1))
```

✓
0
giây

```
[8] xtest=test.iloc[:,0:50]  
    xtest=np.array(xtest)  
    xtest=np.reshape(xtest,(200,50))
```

✓
0
giây

```
[9] ytest=test.iloc[:,50:51]  
    ytest=np.array(ytest)  
    ytest=np.reshape(ytest,(200,1))
```



Training process

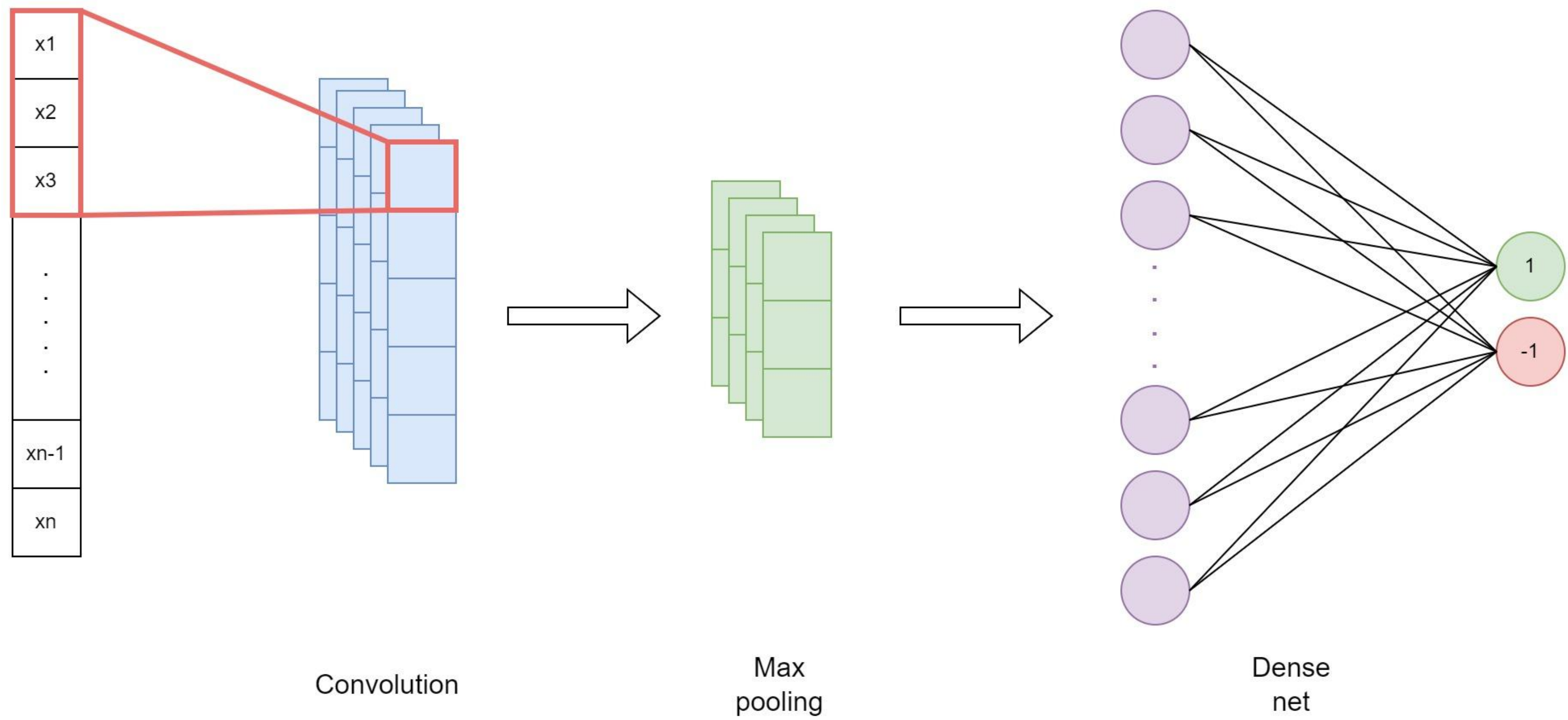
```
model = keras.models.Sequential([
    keras.layers.Conv1D(filters = 32,
                        kernel_size = 2,
                        input_shape = (50,1),
                        activation = "relu"),
    keras.layers.MaxPooling1D(2),
    keras.layers.Flatten(),
    keras.layers.Dense(2, activation='softmax')
])

model.compile(optimizer='adam', loss='mse', metrics=['accuracy'])
model.fit(xtrain, ytrain, epochs=1000)
```

```
Epoch 973/1000
25/25 [=====] - 0s 3ms/step - loss: 1.2755 - accuracy: 0.2375
Epoch 974/1000
25/25 [=====] - 0s 3ms/step - loss: 1.2729 - accuracy: 0.2387
Epoch 975/1000
25/25 [=====] - 0s 3ms/step - loss: 1.2729 - accuracy: 0.2325
Epoch 976/1000
25/25 [=====] - 0s 3ms/step - loss: 1.2732 - accuracy: 0.2400
Epoch 977/1000
25/25 [=====] - 0s 3ms/step - loss: 1.2724 - accuracy: 0.2438
Epoch 978/1000
25/25 [=====] - 0s 3ms/step - loss: 1.2748 - accuracy: 0.2675
Epoch 979/1000
25/25 [=====] - 0s 3ms/step - loss: 1.2742 - accuracy: 0.2525
Epoch 980/1000
25/25 [=====] - 0s 3ms/step - loss: 1.2730 - accuracy: 0.2587
Epoch 981/1000
25/25 [=====] - 0s 3ms/step - loss: 1.2725 - accuracy: 0.2488
Epoch 982/1000
```



Model summary



Giving prediction

```
[[0.6592454  0.34075463]]
```

Evaluate the accuracy

```
[11] test_loss = model.evaluate(xtest, ytest)

7/7 [=====] - 0s 4ms/step - loss: 1.1591 - accuracy: 0.3700
```

Future development

- Collect more data
- Use more features (Open price, volume,...)
- Add more layers into the model
- Use different techniques (LSTM, Random forest,...)
- Try on other stocks (VCB, VIN, ACB,...)



The background is a solid orange color. It features several large, overlapping organic shapes. A large, light blue shape is centered in the middle. To its left, a cream-colored shape overlaps the blue one. To its right, another cream-colored shape overlaps the blue one. In the bottom right corner, a dark green shape is partially visible.

Thank you!

Ask us if u have any question.